

pressure_sensor_noise_test

September 16, 2021

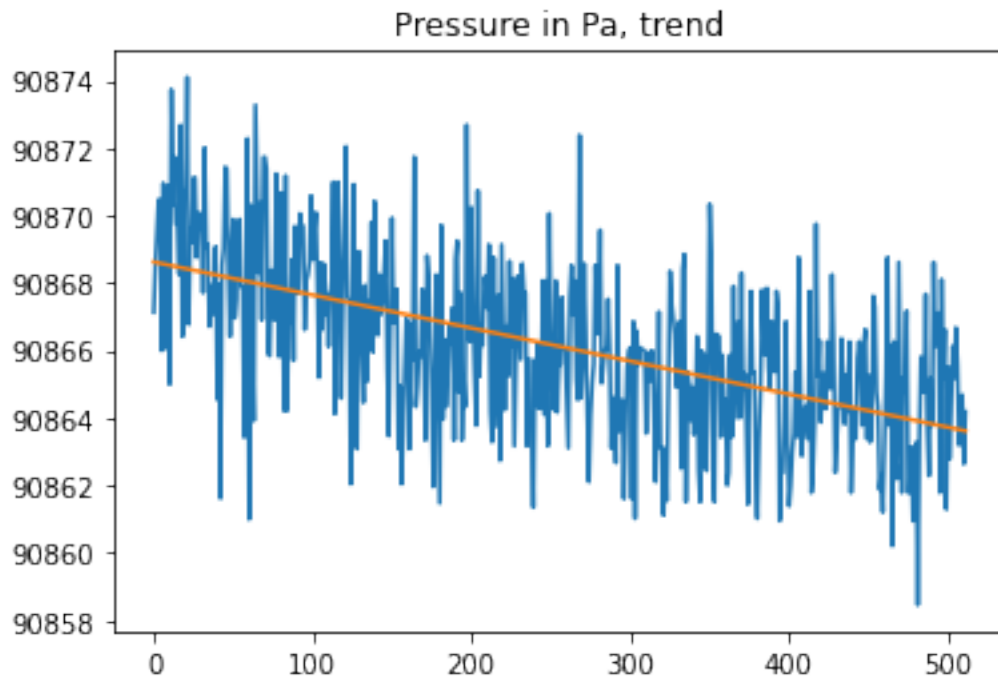
```
[1]: import csv
import statistics
from scipy.stats import normaltest
import numpy as np
from sklearn.linear_model import LinearRegression
import matplotlib.pyplot as plt
from scipy.stats import probplot
```

1 BMP388, 50Hz sample rate (8x pressure, 1x temperature, IIR disabled)

```
[2]: pa = []
zcm = []
with open('bmp388_512samples_1.txt', newline='') as csvfile:
    csvreader = csv.reader(csvfile, delimiter=' ')
    for row in csvreader:
        pa.append(float(row[0]))
        zcm.append(float(row[1]))
pa_arr = np.array(pa)
zcm_arr = np.array(zcm)
pa = []
zcm = []
```

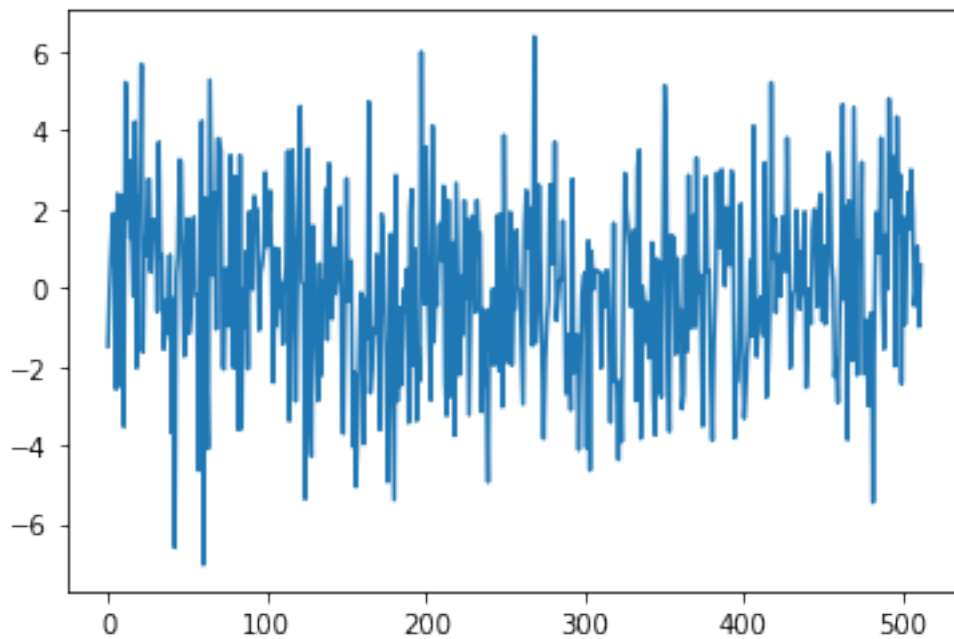
2 Pressure Statistics

```
[3]: X = list(range(0, len(pa_arr) ))
X = np.reshape(X, (len(X), 1))
model = LinearRegression()
model.fit(X, pa_arr)
trend = model.predict(X)
# plot trend
plt.plot(pa_arr)
plt.title('Pressure in Pa, trend')
plt.plot(trend)
plt.show()
```



3 De-trend data

```
[4]: pa_detrended = pa_arr - trend  
plt.plot(pa_detrended)  
plt.show()
```



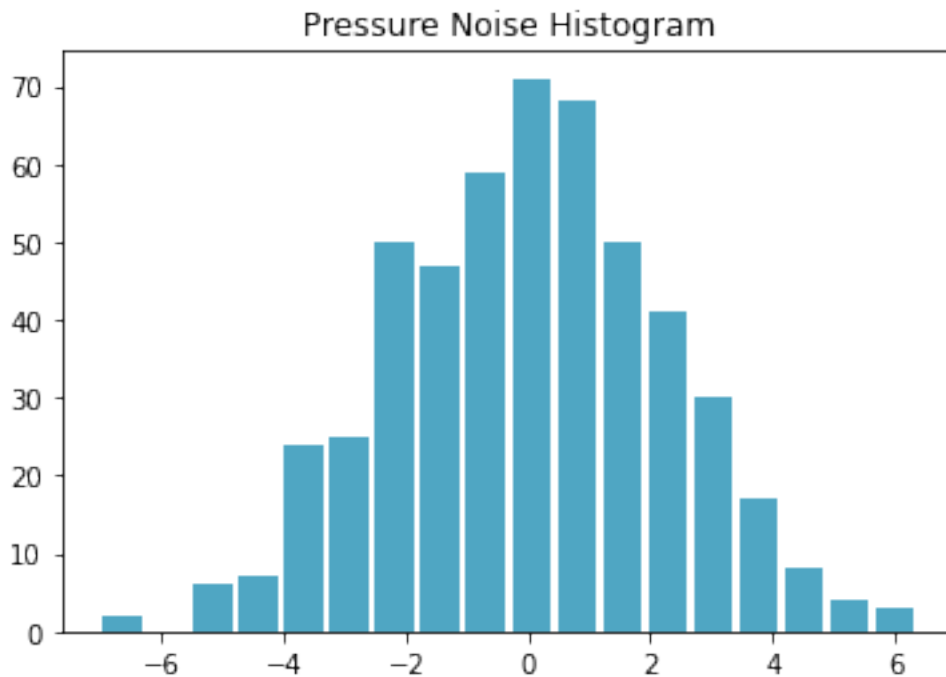
```
[5]: pa_variance = statistics.variance(pa_detrended)
print("Pressure noise variance = {0} Pa^2".format(pa_variance))
```

Pressure noise variance = 4.974331442803269 Pa²

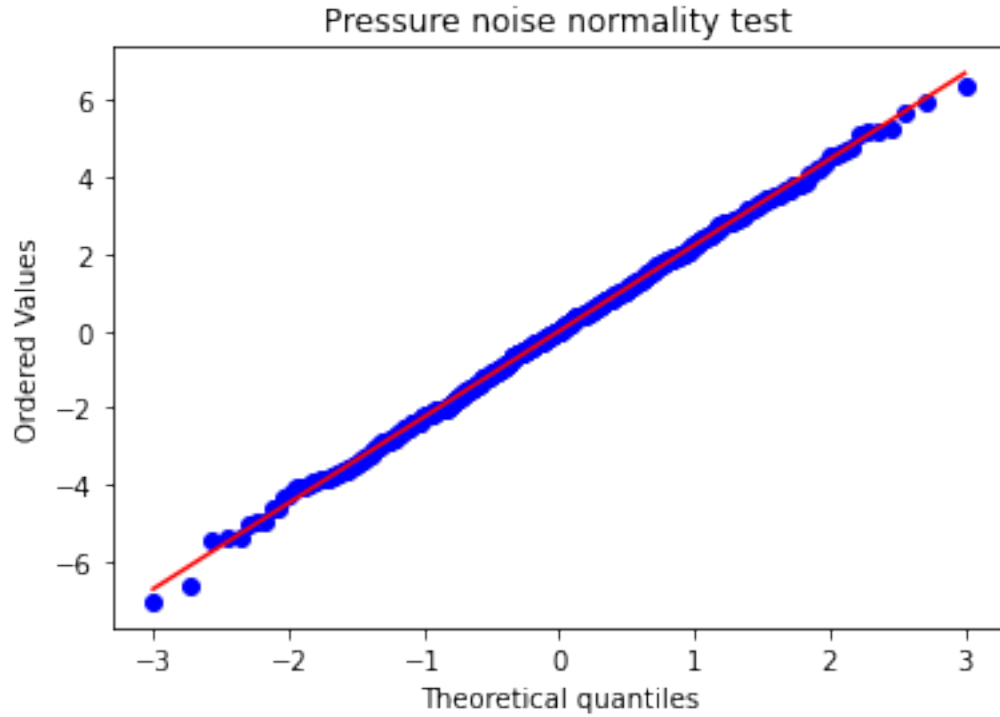
```
[6]: statistic, p_value = normaltest(pa_detrended)
print("Pressure noise normality test : p-value = {0}".format(p_value))
```

Pressure noise normality test : p-value = 0.8529853160704518

```
[7]: n, bins, patches = plt.hist(x=pa_detrended, bins='auto', color='#0580aa',
                                alpha=0.7, rwidth=0.85)
plt.title('Pressure Noise Histogram')
plt.show()
```



```
[8]: probplot(x=pa_detrended,dist='norm',plot=plt)
plt.title('Pressure noise normality test')
plt.show()
```



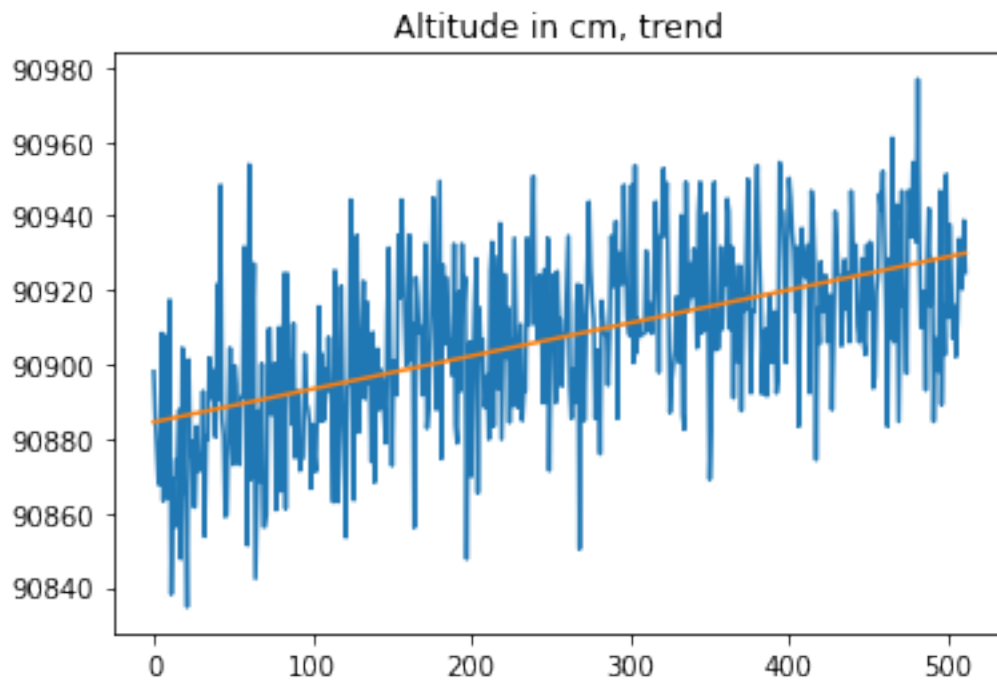
4 Altitude Noise Statistics

On a macro scale, the pressure - altitude relationship is highly non-linear

Pressure Altitude (in cm) = $4430769.396f * (1.0f - \text{pow}(\text{pa}/101325.0f, 0.190284f))$

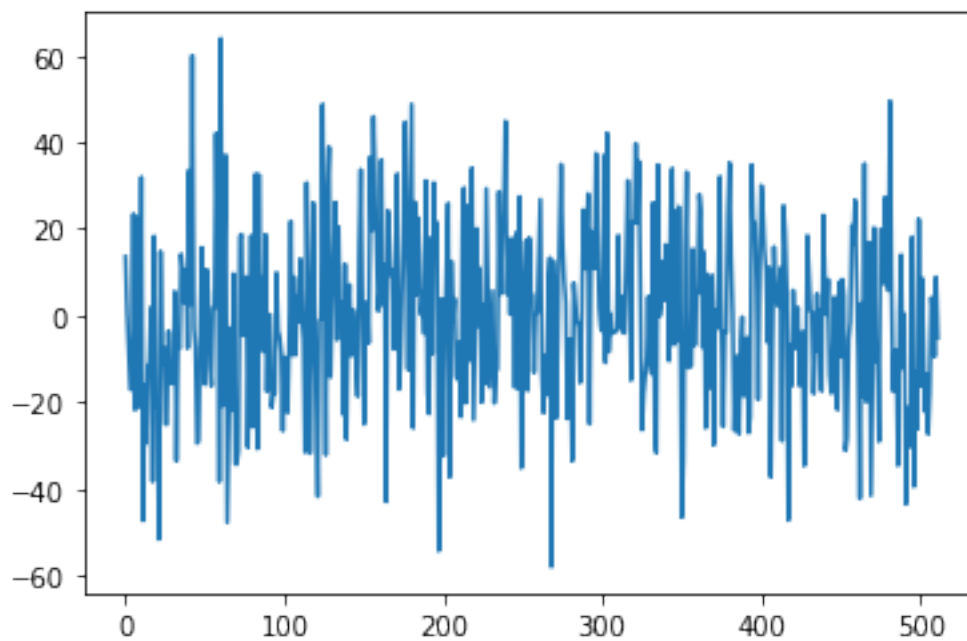
where the pressure is in Pascals (Pa) and sea-level pressure is nominally 101325Pa. However, when you consider a neighbourhood of ± 50 Pascals, the relationship is linear enough so that altitude measurements from a barometric pressure sensor also have a Gaussian distribution for noise.

```
[9]: model = LinearRegression()
model.fit(X, zcm_arr)
trend = model.predict(X)
# plot trend
plt.plot(zcm_arr)
plt.title('Altitude in cm, trend')
plt.plot(trend)
plt.show()
```



5 De-trend data

```
[10]: zcm_detrended = zcm_arr - trend  
plt.plot(zcm_detrended)  
plt.show()
```



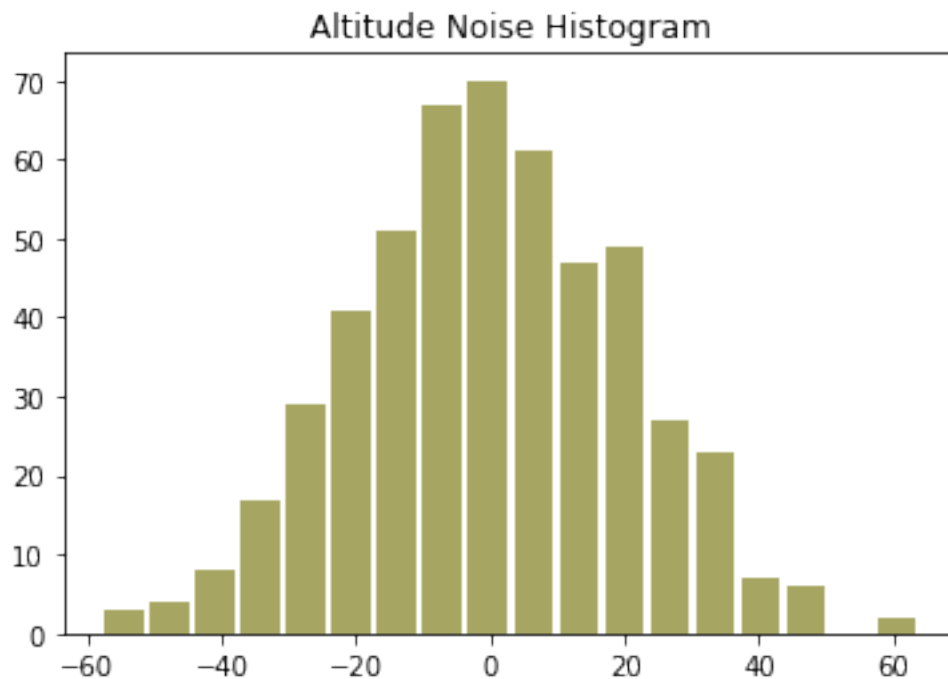
```
[11]: zcm_variance = statistics.variance(zcm_detrended)
print("Altitude noise variance : {0} cm^2".format(zcm_variance))
```

Altitude noise variance : 410.67904353165665 cm²

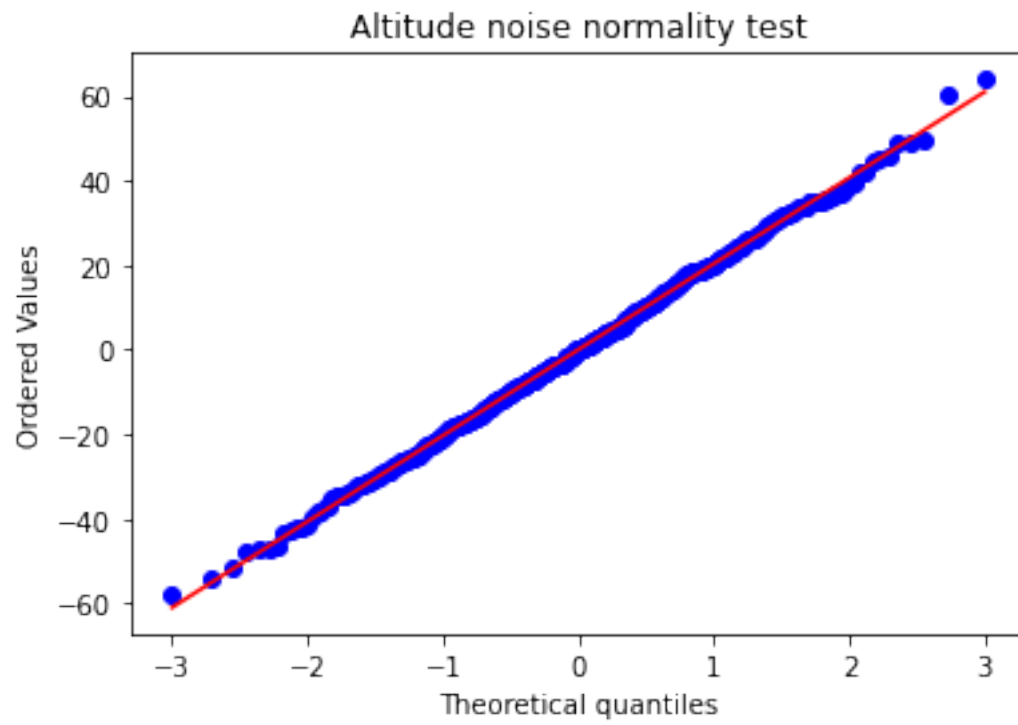
```
[12]: statistic, p_value = normaltest(zcm_detrended)
print("Altitude noise normality test : p-value = {0}".format(p_value))
```

Altitude noise normality test : p-value = 0.8503474135120263

```
[13]: n, bins, patches = plt.hist(x=zcm_detrended, bins='auto', color='#808020',
                                alpha=0.7, rwidth=0.85)
plt.title('Altitude Noise Histogram')
plt.show()
```



```
[14]: probplot(x=zcm_detrended,dist='norm',plot=plt)
plt.title('Altitude noise normality test')
plt.show()
```



[]: